Rotman

A QUICK INTRO TO R R Workshop (M/E MBA)

July 20, 2021 Prepared by Jay Cao / TDMDAL

Website: https://tdmdal.github.io/r-workshop-2021-memba/



Goal for Today – Answer Three Questions

- What's R?
- What can I use R for?
- How to learn R (on my own)?

Plan for Today

- Intro to Intro
 - What is R and what can R do?
 - Setup R
 - Motivation examples
- How to learn R and a quick walk-through
 - Basics of R programming
 - Data science with R
- Learning Road Map and Resources

What's R?



- R = a language + an eco-system
 - A free and open-source programming language
 - An eco-system of many high-quality user-contributed libraries/packages
- In the past R is mostly known for its statistical analysis toolkits
- Nowadays R is capable of (and very good at) many other tasks
 - Tools that cover the whole data analysis workflow
 - Tools for web technology...



What can R do – Statistics & related

- Statistics & Econometrics
 - Regressions
 - Time series analysis
 - Bayesian inference
 - Survival analysis
 - ...
- Numerical Mathematics
 - Optimization
 - Solver
 - Differential equations
 - \bullet ...

- Finance
 - Portfolio management
 - Risk management
 - Option pricing
 - ...
- . . .

See more R Empirical Finance Packages on R Task View - Finance

What can R do – Graphics (static ones)







FANG Candlestick Chart







. . .

<u>https://www.r-graph-gallery.com/</u> https://timogrossenbacher.ch/2016/12/beautiful-thematic-maps-with-ggplot2-only/;

What can R do – Graphics (dynamic ones)



Year: 1952 Africa Americas Asia 80 . 60life expectancy 1e+03 1e+04 1e+05 Europe Oceania 60-40-1e+03 1e+04 1e+05 1e+03 1e+04 1e+05 GDP per capita

https://plot.ly/r/3d-surface-plots/;

https://github.com/thomasp85/gganimate;

What can R do – Others (1)

- Machine learning
 - Statistical learning (clustering, decision tree, etc.)
 - An Introduction to Statistical Learning (with Applications in R)
 - Deep learning (neural networks)
 - Interface to Keras and Tensorflow (via reticulate, an R to Python interface)
 - Torch for R (natively from R; similar as PyTorch in Python)
- Natural language processing (e.g., <u>tidytext</u>, <u>topicmodels</u>)

See more R Machine Learning Packages on <u>R Task View - ML & Statistical Learning</u>
 See more R Natural Language Processing Packages on <u>R Task View - NLP</u>

λα



D Springer



What can R do – Others (2)

- Web technology
 - Web scraping (e.g. <u>rvest</u>)
 - API wrapper (e.g. Twitter: <u>rtweet</u>; bigquery: <u>bigrquery</u>; Quandl: <u>Quandl</u>)
 - Shiny web app (<u>https://shiny.rstudio.com/</u>)
- Reporting
 - <u>R Markdown</u> (write reports, slides, blogs, books, etc. See a gallery <u>here</u>.)
- ... (see <u>**R Task View</u>** for more)</u>

R vs Excel and BI Tools vs Python

- Excel & Business Intelligence (BI) Tools (e.g., Tableau, Power BI, etc.)
 - 2-D tables as basic data structure
 - Good UI (User Interface) and minimum programming
 - Limited modeling tools
 - Not easy to reproduce an analysis (because it's hard to store UI clicks)
 - Not flexible enough for complicated analysis problems, i.e., problems with
 - Many data cleaning steps/pipelines
 - Many different models to try
- Python
 - Very similar to R, but R is more specialized in data analysis
 - R is much easier to learn (in my opinion)



++++ a b | e a u



Why learn R (What can R do for You)?

- Beyond Excel Data Analysis
 - I wish Excel could...
- Automate boring repeating tasks
 - e.g., daily data collection from different sources, weekly dashboard update
- Prototype ideas
 - e.g., a novel trading strategy, a new credit risk model
- Really, find anything that interests you and use R...

Plan for Today

- Intro to Intro
 - What is R and what can R do?
 - Setup R
 - Motivation examples
- How to learn R and a quick walk-through
 - Basics of R programming
 - Data science with R
- Learning Road Map and Resources

Setup R

- R & RStudio on your computer (most of you should choose this one)
 - Install R (<u>https://www.r-project.org/</u>)
 - Install RStudio (<u>https://rstudio.com/products/rstudio/download/</u>)
- R & RStudio in the Cloud (run R without installation)
 - RStudio Cloud (<u>https://rstudio.cloud/</u>)
 - UofT JupyterHub RStudio (<u>https://jupyter.utoronto.ca/hub/login</u>)
- R & Notebook in the Cloud (run R without installation)
 - UofT JupyterHub Notebook (<u>https://jupyter.utoronto.ca/hub/login</u>)
 - Google Colab (<u>https://colab.to/r</u>)



from_col = from,

from_to_map = node_id)

to_col = to,

> g %>% render_graph()

1.1

+

+

>

\times

3

Restudio
File Edit Code View Plots Session Build Debug Profile Tools Help
• • • • • • • • • • • • • • • • • • •
<pre></pre>
<pre>indersy(pig); define the second second</pre>
<pre>1 tbrary(thiggrammek) 5 6 raw <- tribble(7 ~id, ~in_node, ~out_node, ~in_time, ~out_time, 8 #[]] 9 1, 1, 2, 1, 3, 10 1, 2, 3, 3, 5, 11 2, 1, 2, 2, 3, 12 2, 2, 4, 3, 6 13) 14 15 node_tb_tp. <- raw %% 16 distinct(in_node) %% 17 rename(node_id = in_node) 18 19 node_tb <- raw %% 10 distinct(in_node) %% 11 rename(node_id) 18 19 node_tb <- raw %% 10 distinct(in_node) %% 11 rename(node_id) 14 15 arrange(node_id) 15 arrange(node_id) 16 distinct(in_node) %% 17 rename(node_id) 17 rename(node_id) 18 19 node_tb <- raw %% 10 distinct(in_node) %% 11 rename(node_id) 14 15 arrange(node_id) 15 arrange(node_id) 16 distinct(in_node) %% 17 rename(node_id) 17 rename(node_id) 18 arrange(node_id) 19 arrange(node_id) 10 arrange(node_id) 11 arrange(node) %% 11 arrange(node_id) 11 arrange(node) %% 11 arrange(node_id) 11 arrange(node_id) 11 arrange(node) %% 11 arrange(node_id) 11 arrange(node) %% 11 arrange</pre>
6 raw <- tribble(
<pre>7 ~id, ~in_node, ~out_node, ~out_time, 8 #-11-1 9 1, 1, 2, 1, 3, 1, 2, 3, 3, 5, 11 2, 1, 2, 2, 3, 12 2, 2, 4, 3, 6 13) 14 15 node_tb_try < raw %% 16 distinct(in_node) %% 17 rename(node_id = in_node) 18 19 node_tb < raw %% 20 distinct(out_node) %% 21 rename(node_id = out_node) 22 union(node_tb_tp) %% 23 arrange(node_id) 24 25 edge_tb < raw %% 26 distinct(in_node, out_node) %% 27 rename(form = in_node, to = out_node)</pre>
8 #111 9 1, 1, 2, 1, 3, 10 1, 2, 3, 3, 5, 11 2, 1, 2, 2, 3, 12 2, 2, 4, 3, 6 13) 14 15 node_tb_tp <- raw %>% 16 distinct(in_node) %>% 17 rename(node_id = in_node) 18 node_tb <- raw %>% 19 node_tb <- raw %>% 11 distinct(out_node) %>% 12 edge_tb <- raw %>% 13 distinct(in_node, id = out_node) %>% 14 inin(node_tb_tp) %>% 15 node_tb <- raw %>% 16 distinct(in_node, out_node) %>% 17 rename(node_id = out_node) %>% 18 nono(node_tb_tp) %>% 19 node_tb <- raw %>% 19 distinct(in_node, out_node) %>% 20 ininct(in_node, out_node) %>% 21 rename(from = in_node, to = out_node) 22 union(node, to = out_node) 23 arrange (node_id) 24 intinct(in_node, out_node) %>% 26 edge_tb <- raw %>%
<pre>10 1, 2, 3, 3, 5, 10 1, 2, 3, 3, 5, 12 2, 2, 4, 3, 6 13) 14 15 node_tb_tp <- raw %>% 16 distinct(in_node) %>% 17 rename(node_id = in_node) 18 19 node_tb <- raw %>% 20 distinct(out_node) %>% 21 rename(node_id = out_node) %>% 22 union(node_tb_tp) %>% 23 arrange(node_id) 24 25 edge_tb <- raw %>% 26 distinct(in_node, out_node) %>% 27 rename(from = in_node, to = out_node)</pre>
<pre>11 2, 1, 2, 2, 3, 12 2, 2, 4, 3, 6 13) 14 node_tb_tp <- raw %>% 16 distinct(in_node) %>% 17 rename(node_id = in_node) 18 19 node_tb <- raw %>% 20 distinct(out_node) %>% 21 rename(node_id = out_node) %>% 22 union(node_tb_tp) %>% 23 arrange(node_id) 24 25 edge_tb <- raw %>% 26 distinct(in_node, out_node) %>% 27 rename(from = in_node, to = out_node)</pre>
<pre>12 2, 2, 4, 3, 6 13) 14 15 node_tb_tp <- raw %>% 16 distinct(in_node) %>% 17 rename(node_id = in_node) 18 19 node_tb <- raw %>% 20 distinct(out_node) %>% 21 rename(node_id = out_node) %>% 21 rename(node_id = out_node) %>% 22 union(node_tb_tp) %>% 23 arrange(node_id) 24 24 25 edge_tb <- raw %>% 26 distinct(in_node, out_node) %>% 27 rename(from = in_node, to = out_node) 24 24 25 edge_tb <- raw %>% 26 distinct(in_node, to = out_node) 27 27 rename(from = in_node, to = out_node) 24 25 edge_tb <- raw %>% 26 distinct(in_node, to = out_node)</pre>
<pre>13 / 14 15 node_tb_tp <- raw %>% 16 distinct(in_node) %>% 17 rename(node_id = in_node) 18 19 node_tb <- raw %>% 20 distinct(out_node) %>% 21 rename(node_id = out_node) %>% 21 rename(node_id = out_node) %>% 22 union(node_tb_tp) %>% 23 arrange(node_id) 24 24 25 edge_tb <- raw %>% 26 distinct(in_node, out_node) %>% 27 rename(from = in_node, to = out_node)</pre>
<pre>node_tb_tp <- raw %>% distinct(in_node) %>% rename(node_id = in_node) Files Plots Packages Help Viewer distinct(out_node) %>% rename(node_id = out_node) %>% in oncode_tb_tp) %>% arrange(node_id) edge_tb <- raw %>% distinct(in_node, out_node) %>% rename(from = in_node, to = out_node)</pre>
<pre>16 distinct(in_node) %>% 17 rename(node_id = in_node) 18 19 node_tb <- raw %>% 20 distinct(out_node) %>% 21 rename(node_id = out_node) %>% 22 union(node_tb_p) %>% 23 arrange(node_id) 24 24 25 edge_tb <- raw %>% 26 distinct(in_node, out_node) %>% 27 rename(from = in_node, to = out_node) 24 25 edge_tb <- raw %>% 26 distinct(in_node, to = out_node) 27 rename(from = in_node, to = out_node) 28 29 20 20 20 20 20 20 20 20 20 20 20 20 20</pre>
<pre>17 rename(node_id = in_node) 18 19 node_tb <- raw %>% 20 distinct(out_node) %>% 21 rename(node_id = out_node) %>% 22 union(node_tb_tp) %>% 23 arrange(node_id) 24 24 25 edge_tb <- raw %>% 26 distinct(in_node, out_node) %>% 27 rename(from = in_node, to = out_node)</pre>
<pre>i i o node_tb <- raw %>% distinct(out_node) %>% rename(node_id = out_node) %>% union(node_tb_tp) %>% arrange(node_id) edge_tb <- raw %>% distinct(in_node, out_node) %>% rename(from = in_node, to = out_node) %>% rename(from = in_node, to = out_node) %>% </pre>
20 distinct(out_node) %>% 21 rename(node_id = out_node) %>% 21 union(node_tb_tp) %>% 22 union(node_tb_tp) %>% 23 arrange(node_id) 24 25 edge_tb <- raw %>% 26 distinct(in_node, out_node) %>% 27 rename(from = in_node, to = out_node)
<pre>21 rename(node_id = out_node) %>% 22 union(node_tb_tp) %>% 23 arrange(node_id) 24 25 edge_tb <- raw %>% 26 distinct(in_node, out_node) %>% 27 rename(from = in_node, to = out_node)</pre>
<pre>22 unron(node_tb_tp) %>% 23 arrange(node_id) 24 25 edge_tb <- raw %>% 26 distinct(in_node, out_node) %>% 27 rename(from = in_node, to = out_node) 27</pre>
<pre>23 arrange(noue_nd) 24 25 edge_tb <- raw %>% 26 distinct(in_node, out_node) %>% 27 rename(from = in_node, to = out_node) 27</pre>
<pre>25 edge_tb <- raw %>% 26 distinct(in_node, out_node) %>% 27 rename(from = in_node, to = out_node) </pre>
<pre>26 distinct(in_node, out_node) %>% 27 rename(from = in_node, to = out_node) </pre>
<pre>2/ rename(from = in_node, to = out_node)</pre>
29 g <- create_graph() %>%
30
22:24 (Top Level) \$
-/OneDrive/rotman/work/mdl/seelab/graph/ A
>
> edge_tb <- raw %>%
+ distinct(in_node, out_node) %>%
<pre>> rename(irom = in_node, to = out_node) ></pre>
> g <- create_graph() %>%
+ add_nodes_from_table(table = node_tb) %>%
+ add_edges_trom_table(

 \mathbf{v}

RStudio Cloud



RStudio at UofT Jupyterhub

& Science's new Computational and Data Science



R Notebook in Google Colab

🜼 rn1 A Simple Regression - Colal	× +				-	
$ ightarrow$ C' $\mathbf{\hat{c}}$	🛈 🖴 https://colab.research.google.com/github/tdmdal/r-workshop-researchers/blob/master/docs/rn1_A_Simple_Regression.ipynt	⊌ ☆		$\mathbf{\overline{\tau}}$	III\ 🗊	٢
Orn1 A Simple Regr	ession 🗟 Runtime Tools Help			⊂⊃ Sh	are 👗	J
Code 🕂 Text 🛛 💩 Copy to [)rive		ARAM Disk	1 -	🗡 Editin	g 🖍
, 1. Data Import	and Manipulation		1	V ⊕	/ 1	:
We first import a dataset from throughout Wooldridge's text for a simple regression.	the workshop website. This is a dataset on married women labor force participation used in <u>Mroz 1987</u> . The dataset is also used book: Introductory Econometrics: A Modern Approach. After briefly inspecting the data, we create a new column lwage in preparation					
[] # load data data_url <- "https mroz_1987 <- read.	://tdmdal.github.io/r-workshop-researchers/data/mroz_1987.csv" csv(data_url)					
[] # take a look at t str(mroz_1987)	he structure of the data					
See a description of the data	columns <u>here</u> .					
<pre>[] # print the first head(mroz_1987)</pre>	few rows of the dataset					
<pre>[] # create log wage mroz_1987["lwage"]</pre>	<- log(mroz_1987["wage"])					
 Description We will run a simple regression 	n to investigate return on education for married women: $log(wage)=eta_0+eta_1educ+u$.					
[] # setup a regressi lr <- lm(formula =	on model lwage ~ educ, data = mroz_1987)					

R Notebook at UofT Jupyterhub

& Science's new Computational and Data Science



Plan for Today

- Intro to Intro
 - What is R and what can R do?
 - Setup R
 - Motivation examples
- How to learn R and a quick walk-through
 - Basics of R programming
 - Data science with R
- Learning Road Map and Resources

A Few Examples

- Analyze portfolio performance
- Look for trends in R community through Twitter
- Recognize handwritten digits an example of deep learning



A Few Examples: What to Look For

- Focus on analysis workflow (by reading the code comments)
 - Import and manipulate data
 - Model data
 - Report and visualize results
- Don't focus on R syntax
- Do notice everything is done in a sequential way
 - no conditional branching or looping

Plan for Today

- Intro to Intro
- How to learn R and a quick walk-through
 - Basics of R programming
 - Data science with R
- Learning Road Map and Resources

How to Learn R

- Step 1. Basics of R programming
 - Data and programing structures (learn R syntax; relatively easy)
 - How to turn ideas into code (hard; takes time and practice)
- Step 2. Data Science with R
 - Data wrangling
 - Modeling
 - Reporting and visualization
- A Good Learning Approach
 - Learn underlying principles (e.g., why organize data in a certain way)
 - Learn best practices (e.g., follow a consistent analysis workflow)

Free books: Hands-On Programming with R; R for Data Science





Plan for Today

- Intro to Intro
- How to learn R and a quick walk-through
 - Basics of R programming
 - Expression & Assignment
 - Data Structure
 - Programming Structure (control flow & function)
 - Turn ideas into code
 - Data science with R
- Learning Road Map and Resources

Expression and Assignment

```
# expression
```

```
2 + sqrt(4) + log(exp(2)) + 2^2
```

```
# assignment
```

```
x <- 3
```

```
y <- (pi == 3.14)
```

R Data Structure - Overview

	Homogeneous	Heterogeneous
1-d	Atomic vector	List
2-d	Matrix	Data frame
n-d	Array	

R Data Structure - Overview

	Homogeneous	Heterogeneous
1-d	Atomic vector	List
2-d	Matrix	Data frame
n-d	Array	

Atomic Vectors

create R vectors World! vec_character <- c("Hello,", "World!")</pre> Hello, vec_integer <- c(1L, 2L, 3L)2 3 1 vec double <- c(1.1, 2.2, 3.3)</pre> 1.1 2.2 3.3 vec_logical <- c(TRUE, TRUE, FALSE)</pre> TRUE TRUE FALSE

List

```
# create an R list
l1 <- list(
    1:3,
    "a",
    c(TRUE, FALSE, TRUE),
    c(2.3, 5.9)
    1 2 3 "a" TRUE FALSE TRUE 2.3 5.9
</pre>
```

Data Frame

# create a data frame			_
df1 <- data.frame(X	У	Z
x = 1:3,	1	"a"	1.1
y = letters[1:3],	2	"b"	2.2
z = c(1.1, 2.2, 3.3)	3	"c"	3.3

Data Frame

```
# create a data frame
df1 <- data.frame(
    x = 1:3,
    y = letters[1:3],
    z = c(1.1, 2.2, 3.3)
)</pre>
```

х	У	z
1	"a"	1.1
2	"b"	2.2
3	"c"	3.3

Data Frame

```
# create a data frame
df1 <- data.frame(
    x = 1:3,
    y = letters[1:3],
    z = c(1.1, 2.2, 3.3)
)</pre>
```

x	У	z
1	"a"	1.1
2	"b"	2.2
3	"c"	3.3

A Cousin to Data Frame - Tibble

load tibble library (part of tidyverse lib)
library(tibble)

```
# create a tibble
tb1 <- tibble(
    x = 1:3,
    y = letters[1:3],
    z = c(1.1, 2.2, 3.3)
)</pre>
```

https://r4ds.had.co.nz/tibbles.html#tibbles-vs.data.frame

x	У	z
1	"a"	1.1
2	"b"	2.2
3	"c"	3.3

Programming Structure: Control Flows

Today



Learn yourself later (See Appendix)

Sequential

• Example: Sum of Squares



sum of squares
t <- 1:3
y <- sum(t^2)
print(y)</pre>

Sequential

• Example: Sum of Squares



sum of squares
t <- 1:3
y <- sum(t^2)
print(y)</pre>

Sequential

• Example: Sum of Squares



sum of squares t <- 1:3 y <- sum(t^2)</pre> print(y) 1 2 t 3 1 4 9 t^2 sum(t^2) 14

Programming Structure: Functions

- What's a function
 - a logical block of code
 - input -> output
- Why write functions
 - Reusability
 - Abstraction
 - Maintainability

• Example: $\sum_{t=1}^{n} t^2$

sum of squares from 1 to n ss <- function(n) {</pre> t <- 1:n $sum(t^2)$ # calling the ss() function print(ss(2)) print(ss(3))

Programming Structure: Functions

- What's a function
 - a logical block of code
 - input -> output
- Why write functions
 - Reusability
 - Abstraction
 - Maintainability

• Example: $\sum_{t=1}^{n} t^2$

sum of squares from 1 to n ss <- function(n) {</pre> t <- 1:n $sum(t^2)$ # calling the ss() function print(ss(2)) print(ss(3))

Programming Structure: Functions

- What's a function
 - a logical block of code
 - input -> output
- Why write functions
 - Reusability
 - Abstraction
 - Maintainability

• Example: $\sum_{t=1}^{n} t^2$

sum of squares from 1 to n ss <- function(n) {</pre> t <- 1:n sum(t^2) # return(sum(t^2)) } # calling the ss() function print(ss(2)) print(ss(3))

Turn Ideas into Code

- Solve problems using code: three main ingredients
 - Data & Programming Structure + Algorithm (sorting, searching, optimization, etc.)
- Examples
 - Sort a list of integers
 - Generate and solve Sudoku puzzles
 - Implement and backtest a trading rule/algorithm
- For us, in most cases, we solve problems by
 - using other people's algorithm implementations (functions from R packages)
 - Combine algorithms (and data & programming structures) to achieve our goal (still not easy; need practices to write good code.)

Plan for Today

- Intro to Intro
- How to learn R and a quick walk-through
 - Basics of R programming
 - Data science with R
 - A Typical data analysis workflow
 - Choice of R packages
 - An example: regression analysis
- Learning Road Map and Resources

Data Science/Analysis Workflow

• Use this workflow to organize your thoughts and code



Program

An Example: Housing Price & Clean Air

- Manipulate data
 - Load data
 - Create new columns
 - Filter columns and rows
- Build models
 - Multiple linear regressions

Obs: 506

1.	price	median housing price, \$
2.	crime	crimes committed per capita
3.	nox	nitrous oxide, parts per 100 mill.
4.	rooms	avg number of rooms per house
5.	dist	weighted dist. to 5 employ centers
6.	radial	accessibiliy index to radial hghwys
7.	proptax	property tax per \$1000
8.	stratio	average student-teacher ratio
9.	lowstat	% of people 'lower status'

- Report and graph
 - Build a publication-ready table for regression results

R Packages: Many choices, which one to use

- Often, a task can be achieved using functions in different libraries
 - R is open and extensible!
- Example: load a csv file to a data frame
 - Use <u>read.csv()</u> function from the utils library in Base R
 - Use <u>read csv()</u> function from the <u>readr</u> library
 - Use <u>fread()</u> function from the <u>data.table</u> library
 - Use <u>vroom()</u> from the <u>vroom</u> library

R Packages: Many choices, which one to use

- Start with the one most people use
- Choose one that is well maintained
 - check document, github, etc. for last update date
 - packages maintained by companies (e.g., RStudio Co.) or academic teams
- Choose one that suits your task

Great Choice for Data Science Work

• <u>Tidyverse</u>

- "an opinionated collection of R packages designed for data science"
- "All packages share an underlying design philosophy, grammar, and data structures."
- Handle data manipulation, visualization, and much more
- an eco-system: many package developers started to follow tidyverse principles too

• <u>Tidymodels</u>

- "a collection of packages for modeling and machine learning using <u>tidyverse</u> principles"
- Manage modeling process but does not do modeling itself



recipes

tidymodels

parsnip

rsample

Our Choice: the Regression Example

- Manipulate data (<u>tidyverse</u> eco-system)
 - Load data (<u>read csv()</u> from the <u>readr</u>)
 - Create new columns (<u>mutate()</u> from <u>dplyr</u>)
 - Filter columns and rows (<u>select()</u> and <u>filter()</u> from <u>dplyr</u>)
- Build models
 - Multiple regression (<u>lm()</u> from stats library in R base)
- Report and graph
 - Build a publication-ready table (<u>huxreg()</u> from <u>huxtable</u> library)

Using R packages/libraries

• Install an R library (only need to install a library once)

```
install.packages("Library_name")
```

• Load an R library (before you use a library)

library(library_name)

- <u>CRAN</u> (The Comprehensive R Archive Network)
 - <u>CRAN Task Views</u>

Load a CSV file

• <u>read csv()</u> from the <u>readr</u>

read_csv(file)

e.g. hprice <- read_csv("hprice.csv")</pre>

- More about <u>read csv()</u>
- More about <u>readr</u>

Data Manipulation: <u>dplyr</u> basics

• Filter observations (rows): filter()

filter(my_dataframe, condition1, ...)
e.g., hprice_reg <- filter(hprice, price > 20000)

• Select variables (columns): select()

mutate(my_dataframe, new_var1 = expression1, ...)
e.g., hprice_reg <- mutate(hprice_reg, lprice = log(price))</pre>

Create new variables: <u>mutate()</u>

select(my_dataframe, var1, ...)
e.g., hprice_reg <- select(hprice_reg, lprice, rooms)</pre>

Data Manipulation: Data Pipe (%>%)

hprice_reg <- filter(hprice, price > 20000)
hprice_reg <- mutate(hprice_reg, lprice = log(price))
hprice_reg <- select(hprice_reg, lprice, rooms)</pre>

```
hprice_reg <- hprice %>%
filter(price > 20000) %>%
mutate(lprice = log(price)) %>%
select(lprice, rooms)
```

Regression

• Multiple regressions: lm() from stats library in base R

my_model <- lm(y ~ x1 + x2, data)
$$Y_i = \beta_0 + \beta_1 X_1 + \beta_2 X_2 + \epsilon_i$$

my_model <- lm(y ~ x1 + x2 + I(x1 * x2), data) $Y_i = \beta_0 + \beta_1 X_1 + \beta_2 X_2 + \beta_3 X_1 X_2 + \epsilon_i$

Regression result summary: summary()

Ref. https://faculty.chicagobooth.edu/richard.hahn/teaching/FormulaNotation.pdf

Report

- Summary table
 - <u>Summary for Im()</u>: summary(my_model)
- publication-ready table: <u>huxreg()</u> from <u>huxtable</u> library

huxtable(my_model1, my_model2, ...)

Ref. https://hughjonesd.github.io/huxtable/huxreg.html

Plan for Today

- Intro to Intro
- How to learn R and a quick walk-through
 - Basics of R programming
 - Data science with R
- Learning Road Map and Resources

Learning Road Map (Three Free Books)

- Step 1. Basic R programming skills (Never programmed before? Start here.)
 - Data and programming structure; how to turn an idea into code;
 - Book: <u>Hands-On Programming with R</u>
- Step 2. R Data Science skills
 - Data wrangling, modeling, and visualization/reporting; Best practice;
 - Book: <u>R for Data Science</u>
- Step 3. Take your R Skill to the next level
 - Book: <u>Advanced R</u>

Other free books check **bookdown.org** often



Free Learning Resource

- <u>RStudio Education</u>
 - <u>Choose Your Learning Paths</u>
- <u>RStudio Video Resources Site</u>
- More free R books? Check <u>bookdown.org</u> often
- Coursera: Search R and learn
 - free for <u>UofT students</u> (mostly always free if you just audit the courses)
- Twitter (a few seeds: <u>#rstat</u>, <u>@hadleywickham</u>, <u>@WeAreRLadies</u>)

Appendix

• Programming Structure Continued

- Conditional
- Iteration

Conditional (if...else...)

```
if (cond) {
    # run here if cond is TRUE
} else {
    # run here if cond is FALSE
```

```
# y greater than 10?
if (y > 10) {
    print("greater than 10")
} else {
    print("less or equal to 10")
}
```

Conditional (if...else...)

```
if (cond) {
    # run here if cond is TRUE
} else {
    # run here if cond is FALSE
```

```
# y greater than 10?
if (y > 10) {
  print("greater than 10")
} else {
  print("less or equal to 10")
}
                        F
               y>10?
     "great..."
                       "less..."
```

Conditional (if...else if...else...)

- if (cond1) {
 - # run here if cond1 is TRUE
- } else if (cond2) {
 - # run here if cond1 is FALSE but cond2 is TRUE
- } else {
 - # run here if neither cond1 nor cond2 is TRUE

Iteration

for (var in seq) {
 do something
}

while (cond) { do something if cond is TRUE }

```
# sum of squares
t <- 1:3
y <- 0
for (x in t) {
  y < -y + x^{2}
}
print(y)
```